

# A National Web Conference on Improving Health IT Safety Through the Use of Natural Language Processing to Improve Accuracy of EHR Documentation

### Presented by:

Thomas Payne, M.D. Li Zhou, M.D., Ph.D.

### **Moderated by:**

Chris Dymek, Ed.D.
Agency for Healthcare Research and Quality

February 7, 2017



## **Agenda**

- Welcome and Introductions
- Presentations
- Q&A Session With Presenters
- Instructions for Obtaining CME Credits

**Note:** After today's Webinar, a copy of the slides will be emailed to all participants.



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To produce evidence to make health care safer, higher quality, more accessible, equitable, and affordable, and work within the U.S. Department of Health and Human Services and with other partners to make sure that the evidence is understood and used.



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  - Utilizing Health Information Technology to Scale and Spread Successful Practice Models Using Patient-reported Outcomes
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# Presenter and Moderator Disclosures

The following presenters and moderator have no financial interests to disclose:

- Thomas Payne, M.D.
- Li Zhou, M.D., Ph.D.
- Chris Dymek, Ed.D.

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## **Presenter Grant Support**

### Recent grant support for our presenters includes:

Dr. Thomas Payne: Dr. Li Zhou

AHRQ HS023631 AUR A100077 AHRQ HS024264 AHRQ HS022728

Controlled Risk Insurance

Company (CRICO)

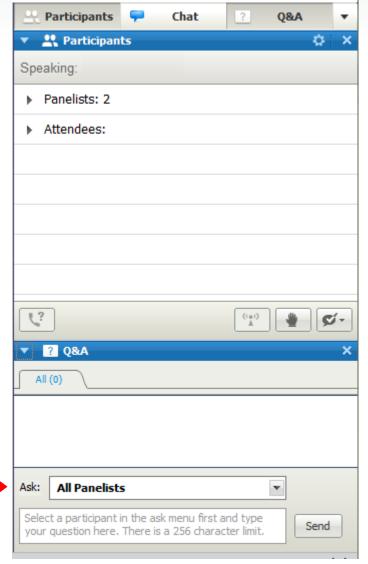
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- Please address your questions to "All Panelists" in the drop-down menu.
- Select "Send" to submit your question to the moderator.
- Questions will be read aloud by the moderator.





## **Learning Objectives**

At the conclusion of this activity, the participant will be able to do the following:

- 1) Discuss the development and evaluation of an enhanced electronic note system that leverages voice recognition and Natural Language Processing (NLP) technologies to create electronic physician notes in the EHR.
- 2) Discuss the challenges of introducing speech recognition technology into existing medical culture and current clinician workflow, including user preferences and the quality of documents generated by this technology.
- 3) Explain the need for an automated error detection system using NLP for improving the accuracy and quality of speech recognition generated medical documents, and discuss the development and evaluation of such a system.

9





# Improving Accuracy of Electronic Notes Using a Faster, Simpler Approach

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## **Objectives**

- Review problems with current physician notewriting practices.
- See how these problems might be addressed using current technologies and commercial EHRs.
- Understand barriers to changing physician documentation practices and how to address them.



### **Disclaimer**

- Analysis still underway
- Statistical testing not yet performed

Final results may differ from those presented here.



### **AHRO** Terms Used in This Webinar

**Automatic speech recognition:** Using software to convert spoken speech into text.

Ways to use automatic speech recognition software:

- Interactive: The user speaks into a microphone and watches the screen as the voice is converted to text and the user corrects errors interactively.
- Noninteractive: The user creates a voice file containing the
  entire document by speaking into a telephone or voice recorder.
  Automatic speech recognition software converts the voice file to
  text in the background while the user is engaged in other
  activities. A transcriptionist or the user corrects errors.



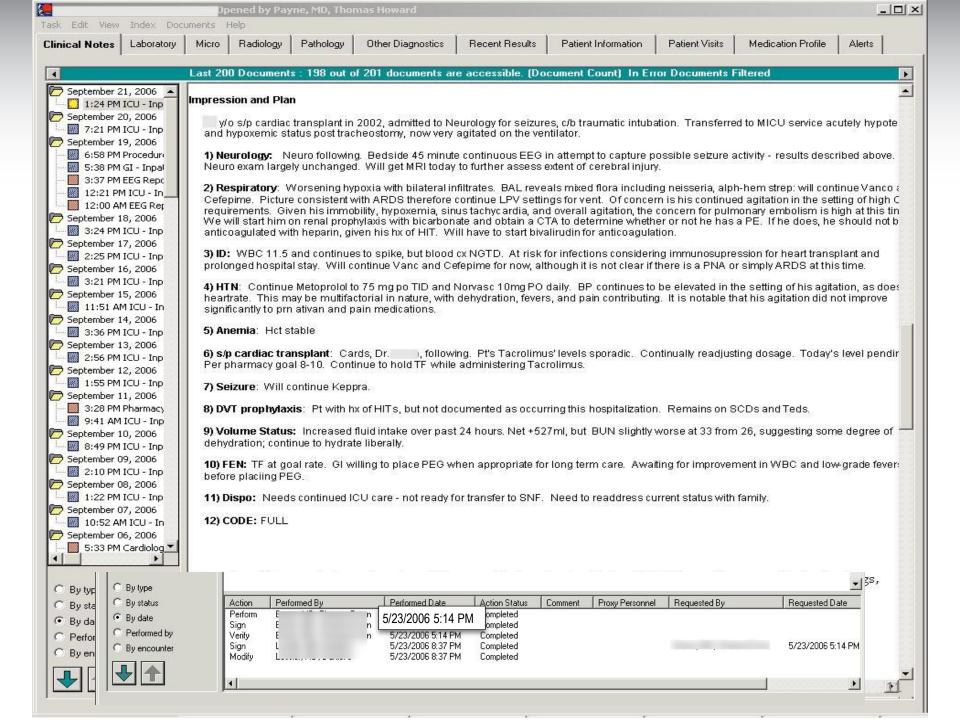
# What Hath We Wrought?





### **AHR** Other Problems With Notes

- Note bloat.
- Copy and paste is common, usually as a result of efforts to save time.
- Progress notes are finished so late that other team members may not see them until the next day.





## **Specific Aims**

Specific Aim 1

To refine and implement a new **voice**-generated enhanced electronic note system (VGEENS), integrating voice recognition and transcription with natural language processing and links to the electronic medical record (EMR) to improve note creation efficiency and note accuracy.

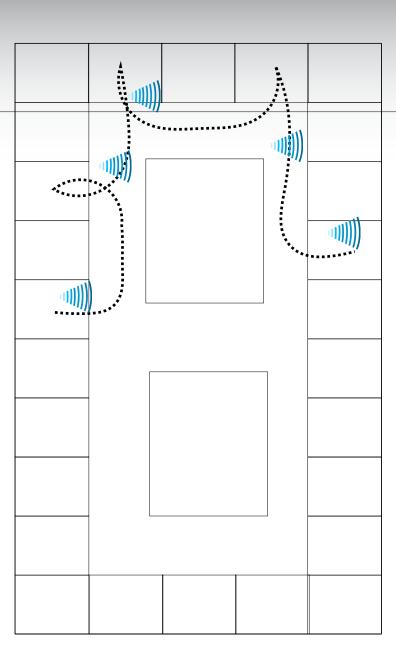
Specific Aim 2

To evaluate VGEENS using a **randomized trial** with 30 internal medicine physicians in each arm to assess electronic note creation efficiency, note accuracy, and user satisfaction. Intervention physicians will use VGEENS, while control physicians will continue with note creation as they normally would.



### Specific Aim 1

Notes will be 'dictated' at the bedside or immediately after leaving it.





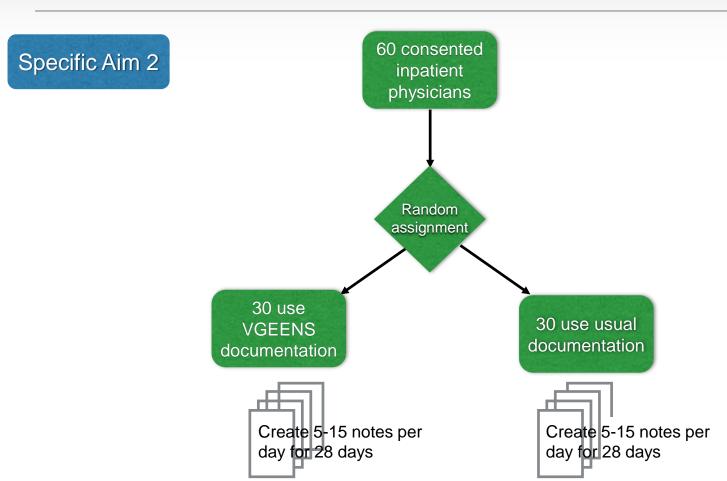
# Ways to Write Inpatient Progress Notes

	Fast	Low cost	Incorporates EMR data	Encoded data	Well suited to rounding workflow
VGEENS	✓	✓	✓	✓	✓
Keyboard		✓	✓	✓	
Dictation	✓				✓
Voice recognition		<b>√</b>			



### **Randomized Controlled Trial**

(Planned)



- 1. Minutes between when patient seen on rounds and note signed in EMR
- 2. Note quality, measured by Physicians Documentation Quality Instrument
- 3. Satisfaction of physician users



# Physicians Documentation Quality Instrument

Attribute	Score					Description of Ideal Note
1. Up-to-date	Not at all	2	3	4	Extremely 5	The note contains the most recent test results and recommendations.
2. Accurate	Not at all	2	3	4	Extremely 5	The note is true. It is free of incorrect information.
3. Thorough	Not at all	2	3	4	Extremely 5	The note is complete and documents all of the issues of importance to the patient.
4. Useful	Not at all	2	3	4	Extremely 5	The note is extremely relevant, providing valuable information and/or analysis.
5. Organized	Not at all	2	3	4	Extremely 5	The note is well-formed and structured in a way that helps the reader understand the patient's clinical course.
6. Comprehensible	Not at all	2	3	4	Extremely 5	The note is clear, without ambiguity or sections that are difficult to understand.
7. Succinct	Not at all	2	3	4	Extremely 5	The note is brief, to the point, and without redundancy.
8. Synthesized	Not at all	2	3	4	Extremely 5	The note reflects the author's understanding of the patient's status and ability to develop a plan of care.
9. Internally Consistent	Not at all	2	3	4	Extremely 5	No part of the note ignores or contradicts any other part.
Total Score:						



# Results



# **VGEENS System Successfully Developed**

### Specific Aim 1

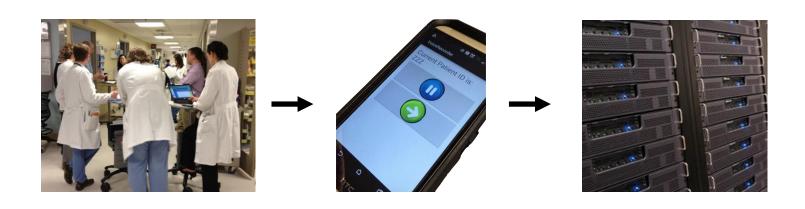
- Used with commercial **EHR**
- Notes available in EHR Inbox within 5 minutes
- Secure
- Enhanced with text processing later in trial
- Downtime uncommon





### Specific Aim 1

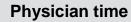




Voice dictation files will then be securely transmitted to servers...



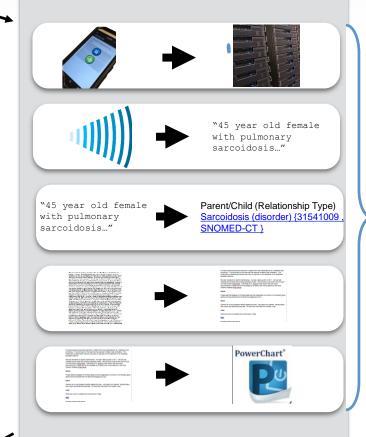
Specific Aim 1



Record note (~5 minutes),

Edit and sign note (~3 min)

### Computer time



5 minutes



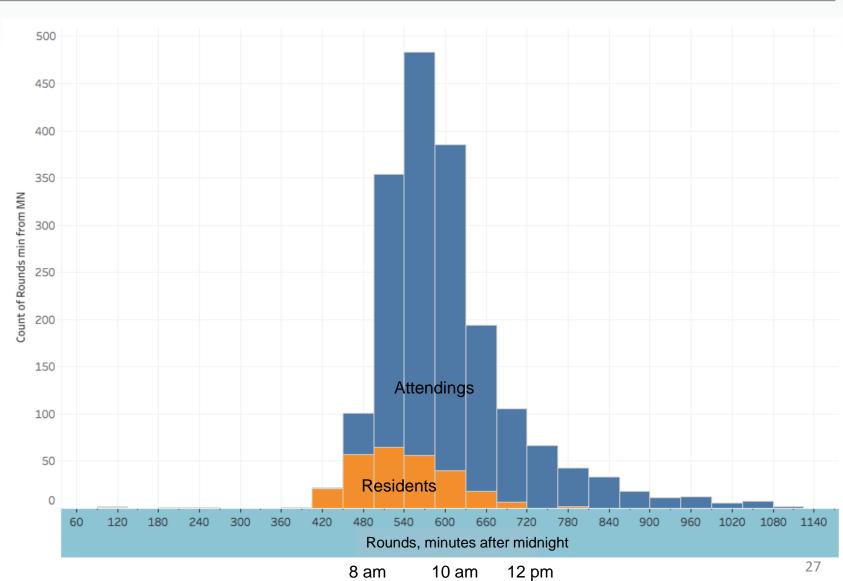
# Summary of Subjects, Outcome Data, and Results

### Specific Aim 2

	Intervention	Control
SUBJECTS		
Consented	24	25
Wrote ≥ 1 note (remainder results from these)	13	18
OUTCOME DATA		
Recorded rounding time (%)	99	99
Notes written	709	1143
Satisfaction survey response (%)	100	100
OUTCOME RESULTS		
Timing: Note available – Rounding time (minutes)	227	190
Satisfied: Highly or moderately (%)	40	50
Dissatisfied: Moderately or dissatisfied (%)	40	6
Note quality: Pending		

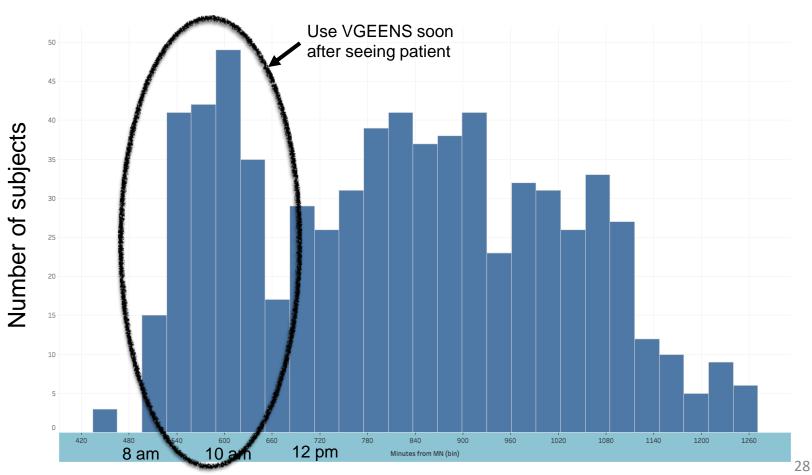


## **Number of Minutes From Midnight** When Patients Seen on Rounds



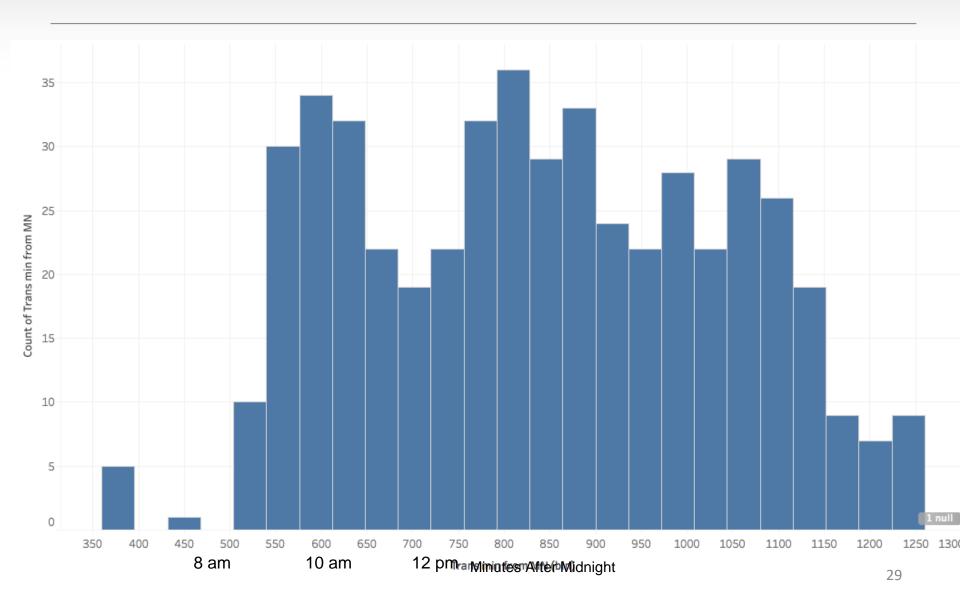


## **INTERVENTION: Number of Minutes** From Midnight When VGEENS Used



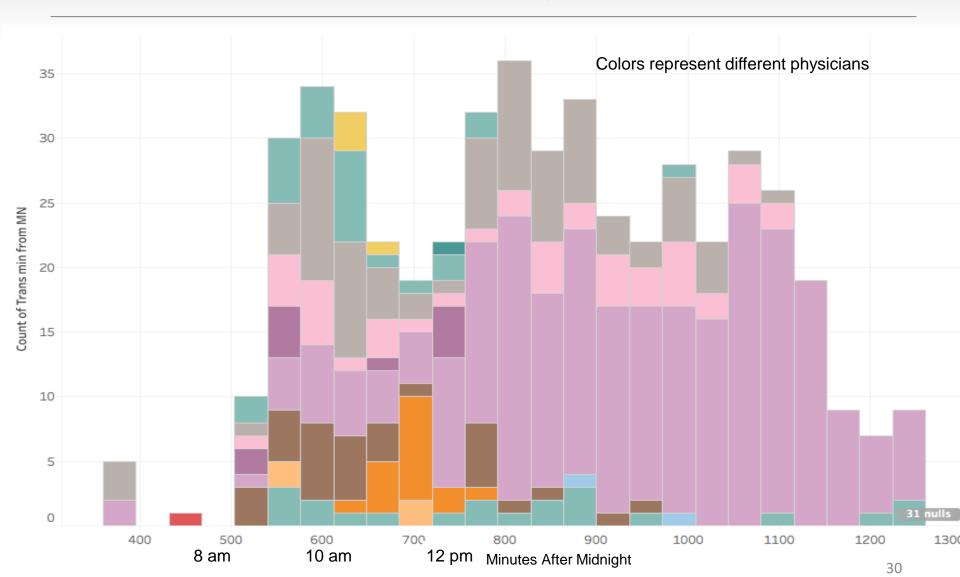


# Number of Minutes From Midnight When VGEENS Notes Available



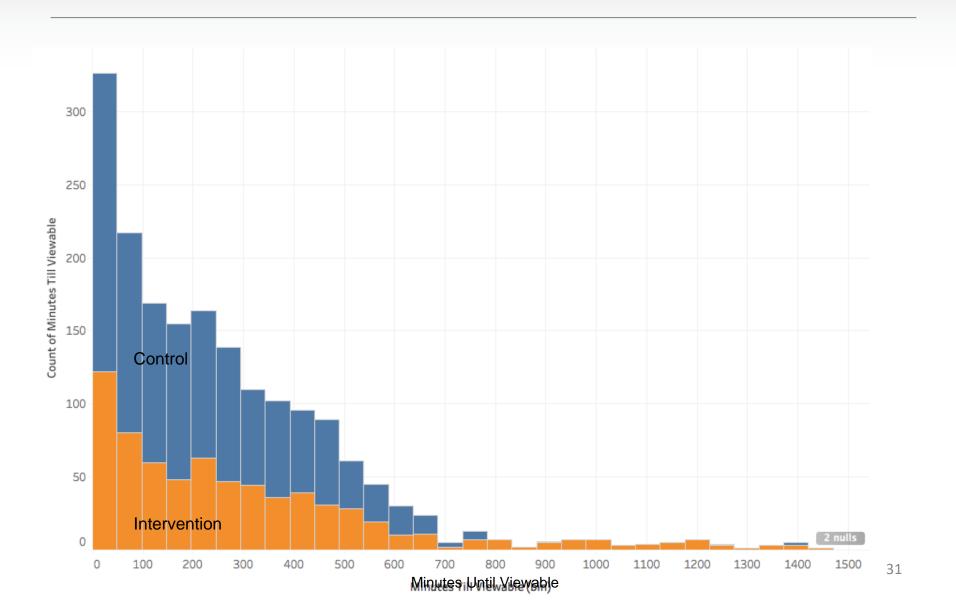


# Minutes After Midnight Notes Transcribed, by Author



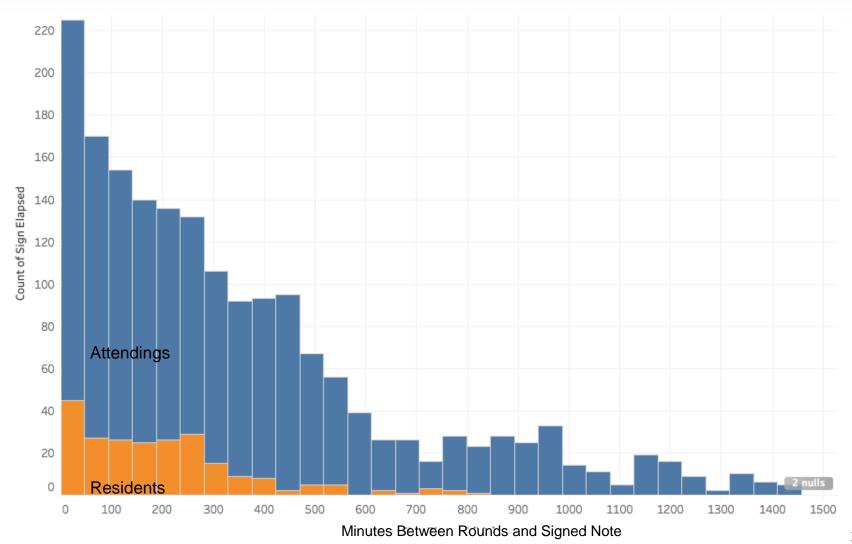


# Time Until Note Viewable Intention to Treat





# Time Between Rounds and Signed Note (Control)





# Comparing VGEENS and Control Notes With Note Written Previous Day on Same Patient

INPATIENT PROGRESS NOTE

### VGEENS

Identification/chief complaint: This is a -year-old woman with metastatic adenocarcinoma of unknown primary mostly involving the hiphead then started and steenum-admitted for pain-Significantly increased pain overnight. Her fentanyl PCA had been reduced from 50 to 25 but she required several subsequent boluses. She is in significant pain this morning and is barely able to speak as a result of that. Her pain is again located with Dr. Carolyn Sw this morning who acted as a second opinion. The social worker for the death w/ dignity program will meet with her as well. Medications were reviewed. For pain, she has an intrathecal pump andwith morphine and bupivacaine. She also is on a fentanyl PCA with an incremental dose of 75 pg Vitals: Temperature 36.2-37.63, heart rate 76-9590, respiratory rate 12-1615-16, oxygen saturation 95-971,981 on room air, blood pressure 88-898/49-8586-98/52-62 General: Thin featiChronically ill-appearing woman lying in bed geimacingin clear Cardiovascular: Regular rate and rhythm Respiratory: Breathing is unlabored, Normal respiratory effort and rate is completely normal on room air, lungs are clear anteriorly Abdomen: Soft, pondistended Neuro: She is fully awakeAlert and oriented, no signs of semsolence <4 This is a -year-old woman with adenocarcinoma of unknown primary admitted for uncontrolled cancer related pain in the left hip and sternum. She remained remains inpatient due to possly conscolled for titration of her pain with a plannedmedications before discharge to inpatient hospice.

1. Acute on chronic pain: He continues to be Her pain was very poorly controlled despite on up titration of her yearnight. Her fentanyl FCAr dose has subsequently been increased back to 75 pg. Her CayContin which had been increased from 120 to 180 mg 3 times a day has been discontinued. I discussed with the patient that it is very likely she will continue to require the FCA at discharge, which she seemed agreeable to. We'll continue to work with the chronic pain service on optimizing her pain management. When she is ready to discharge, we will need to coordinate pain pumpicapylies with the and work with social work on transitioning to impatient hospite serviceys, hose hospice. Of note, the transport from the hospital to hospice may be challenging due to her intrathecal catheter. We are still clarifying the logistics of this.

 Adenoma adenomacerofonem\_Adenocarcinoma of unknown primary: Retient has a peer prognosis with plan dischergethe patient and her family are agreeable to inpatient hospice or home hospice. She has discussed been discussing death with dignity with her outpatient oncologist as well as Dr. Her desphere is due to fit back feen foundary. The says that she does a still want to commit a warke enough to

3. Hypotension: Blood pressures  $\frac{\text{have}}{\text{remained}}$  remained  $\frac{\text{soft but}}{\text{stable}}$  stable after discontinuation of  $\frac{\text{hor}}{\text{lower}}$  IV fluids.

GI/FEN: General diet
Prophylaxis: Low molecular weight heparin
CODE STATUS: DAN/DNI
Disposition: Hospice potentially at home va. impatient Evergreen hospice once her pain regimen has been optimized.

#### Control

THE INPOS





# What Is Medical Informatics?

(attributed to Homer Warner)

Technology 10%
Medicine 10%
Sociology 80%



### **Lessons Learned**

- Physicians may resist changing established note-writing habits.
- The VGEENS approach can make creating progress notes faster if voice recorded at bedside or soon after.
- Notes created using voice may contain less text carried forward from prior notes and may be more accurate.
- Features popular with physicians: carrying forward plan, 'checklist' information, minimizing editing requirements.



### Successes

- We developed and deployed a new system, integrated with a commercial EHR, to create inpatient progress notes within 5 minutes.
- If used at the bedside or soon thereafter, notes are available much sooner for others to view.
- Notes created using voice may contain less text carried forward from prior notes.
- We have a method to apply decision support based on progress note content within minutes.



#### Challenges

- Physicians may resist changing established note-writing habits.
- On average, satisfaction was greater with usual (control) method of note writing, perhaps because popular VGEENS features weren't available until late in the controlled trial.



# Lessons From a First-Time Principal Investigator

- Developing a system to be used in producing and conducting an RCT in 2 years is ambitious, but possible.
- People are enthusiastic about joining an interesting project. Diverse disciplines help!
- Pick a problem you know is important and understand deeply.
- Think about the next project from Day 1 (automated editing, NLP tools, measuring note accuracy, etc.)



#### **Our Team**









#### **Co-investigators**

**Andrew White** Meliha Yetisgen Tom Gallagher

**Computing engineer Andrew Markiel** 

Research team Amelia Chappelle Jennifer Zech~







#### **Collaborators**

David Alonso Xinran (Leo) Liu Ross Lordon Kevin J. Lybarger Mari Ostendorf **Trevor Steinbach** 











#### Thanks to AHRQ!

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# Improving Health IT Safety Through the Use of Natural Language Processing to Improve Accuracy of EHR Documentation

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#### **Document Quality & Patient Safety**

- Accurate medical documents are critical for safe patient care and effective inter-provider communication.
- Errors in medical documents can lead to medical errors in patient care, some of which cause injury or even death.
- ~5 million errors per year are tied to wrong medications;
   1 in 4 medication errors involves a pair of drugs whose names look alike or sound alike.
  - Altenol vs. Atenolol
  - Lyrica vs. Lamictal



#### **PHRO Spelling Errors in Clinical Documents**

- Non-word errors (e.g., Humulog for Humalog)
  - Free-text entries, typed notes
- Real-word errors (i.e., the word is spelled correctly but is contextually wrong, such as there for their)
  - Speech Recognition (SR) generated text



### **Background: Non-Word Errors Detection and Correction**

Spelling errors in free-text EHRs

	Notes (n=315)	Free-text allergy entries (n=2626)	Free-text med orders (n=2743)
Error rate	0.5%	4.5%	7.5%
Clinical terms	28.2%	65.5%	78.0%
Real-word errors	3.8%	1.8%	0%

We developed a spell checker in our MTERMS NLP system

	MTE	ERMS Spell	Aspell Default	
	Notes	Allergies	Medications	Notes
Precision	71.1	96.2	90.0	48.9
Recall	81.0	92.7	91.5	82.3
F measure	75.7	94.4	90.8	61.3
Accuracy	78.1	88.2	81.5	58.5

# Background: Speech Recognition and Real-Word Errors

- Previous studies were limited in scope and sample size
- Error rates by word range from 1.5%-15%
- High error rate by report (23%) were found in radiology reports
- 76% of radiologists believed error rate by report < 10%</li>

Author	Doc Type	Sample Size	Error Rate
Devine*	A discharge summary	12 physicians	IBM software: 7.0% to 9.1%
2000	and a progress note		Dragon: 14.1% to 15.2%
Kanal*	72 radiology reports	6 participants	IBM MedSpeaker: 10.3%
2001			Significant errors: 7.8%
Zick*	47 ED charts	2 physicians	Dragon NaturallySpeaking: 1.5%
2001			Errors/chart: 2.5
Quint	265 radiology reports	-	22% of reports contained significant errors
2008			
Basma	615 radiology reports	-	23% of SR reports contained major errors
2011			4% in conventional dictation transcriptions

<sup>\*</sup> Error Rates: total number of errors divided by the total number of words in the report.

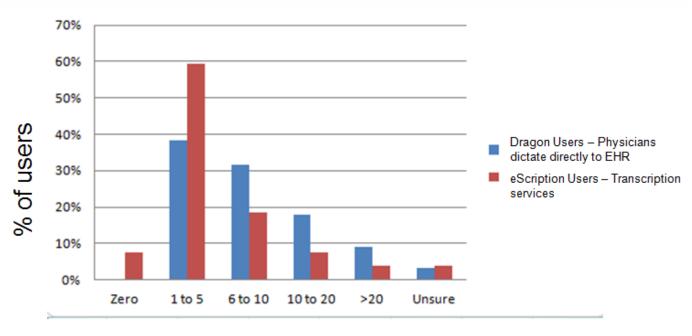


#### **Background: Our SR Error Study in ED**

- Retrospectively analyzed 100 emergency physician notes during Dec 2012.
  - Generated via a front-end SR system (Dragon® Medical 10.0)
  - Further edited and signed by the physicians
- 71% of notes contained errors; 1.3 errors per note; 9 errors per 1000 words.
- 15% contained one or more clinically significant errors.
- Physicians signed their notes with known errors, indicating proofreading the entire medical note to search for errors is time consuming.



#### **AHR** Background: Our Clinician Survey



Number of Errors per Note

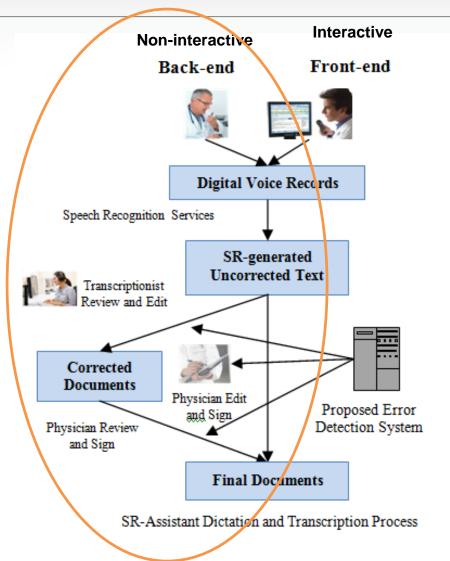
Surveyed 114 Dragon Users and 50 eScription Users at Brigham and Women's Hospital, Boston



#### **SR Error Analysis and Detection**

(Ongoing Study Funded by AHRQ)

- Aim 1: Conduct error analysis to estimate the prevalence and severity of SR errors.
- Aim 2: Develop NLP methods for automated error detection.
- This presentation reports our errors analysis in back-end SR generated documents at different processing stages.





#### **Methods**

- Stratified random sample of 169 dictated notes using transcription services (back-end SR)
  - ▶ 79 from Brigham and Women's Hospital (24 operative notes and 55 office notes)
  - ▶ 40 discharge summaries from North Shore Medical Center, Boston
  - 50 from the University of Colorado Hospital (35 discharge summaries and 15 operative notes)
- Four processing stages
  - Original audio file dictated by the provider (AO note)
  - ► Note generated by SR engine of the vendor transcription service (**SR** note)
  - ► Note edited by a professional medical transcriptionist (**MT** note)
  - Final note reviewed and signed by a clinician (SN note)
- Three-level annotation schema
  - General error types
  - Semantic error types
  - Clinically significant errors
- Manual review to create gold standard



#### **Measures**

- Length of time to dictate a note
- Turnaround time for each note version
- Differences in the SR note, MT note, and SN note from the gold standard
- Error rate: number of errors divided by the number of words
- Percentage of each error type by overall errors
- Percentage of notes with at least one clinically significant error
- Repeated these analyses for SR, MT, and SN notes; for each note type; and across all notes



### **General Error Types**

	Туре	Description	Example	
	Insertion	One or more words was added to the transcription	AO: There is distal biliary obstruction observed SR: There is <b>no</b> distal biliary obstruction observed	
	Deletion	One or more words was deleted from the transcription	AO: CHADS2 VASC score 4 SR: score	
	Enunciation	An error due to a mispronunciation or failure to enunciate on the part of the speaker	AO: to find a homeopathic provider SR: defined homeopathic provider	
	Suffix error	The root word is correct, but there is an incorrect, added, or omitted suffix	AO: mental status worsened SR: mental status worsens	
on	Dictionary error  An error due to the target word not being present in the SR system's dictionary		AO: driving a Camry and hit another car SR: driving an Academy and hit another car	
Substitution	Spelling error  The <b>transcriptionist</b> made a spelling error when editing the output of the SR system		AO: we counseled him on risk of infection MT: we counseled hom on risk of infection	
Su	Homonym error One word has been substituted for another identically pronounced word		AO: serial high resolution anoscopy SR: cereal high resolution anoscopy	
	Nonsense error A substitution that is so far off that it can determined which (if any) category it fa		AO: follow up in 3 to 5 days SR: neck veins are evaluated	
	Prefix Error	The root word is correct, but there is an incorrect, added, or omitted prefix	AO: Inadequate evaluation to exclude neoplasia SR: Adequate evaluation to exclude neoplasia	
	Number error	Any error involving a number, whether it is written as a digit ("3") or as a word ("three")	AO: the patient is a <mark>17</mark> -year-old female SR: the patient is a <mark>70</mark> -year-old female	
	Punctuation Error	A period, comma, or other punctuation mark was added where it should not have been	AO: at discharge she had no flank tenderness SR: at discharge. She had no flank tenderness	



### **Semantic Error Types**

Туре	Description	Example
General English	Any English words that do not fit into the categories below	AO: which she would otherwise forget SR: which she would otherwise for gas that
Stop Word	Common English words (we are using the list defined at <a href="http://ranks.nl/stopwords">http://ranks.nl/stopwords</a> )	AO: intermittent pain under the right breast SR: intermittent pain in the right breast
Medication	Medication names and dose information	AO: initiated on Lamotrigine therapy SR: initiated on layman will try therapy
Diagnosis	Any words that are part of a specific medical diagnosis	AO: Dengue SR: DKA
Lab	Includes lab test names and lab test results	AO: TSH of <mark>26.7</mark> SR: TSH of <mark>22nd 6.7</mark>
Imaging Test	Imaging exam names/types and exam results	AO: nonobstructive on CT imaging SR: nonobstructive on imaging
Procedure	Procedure names and descriptions	AO: CA ligament was released on the leading edge SR: CA ligament was released operating edge
Physical exam	Any information directly related to the physical exam (ht/wt, HR, BP, etc.) and any associated values	AO: T 36.7 degrees SR: T3-T7 disease
Patient/ provider info.	Any words involving patient/provider metadata, such as the patient's name, doctors' names, patient MRN, etc.	AO: SURGEON: [surgeon's actual name] SR: SURGEON: Stathis stairs
Date	Any dates, including those that are written with words (January 1, 2017) or written with numbers (01/01/2017)	AO: 10/10/ <mark>2016</mark> SR: 10/10/ <mark>2000</mark>
Symptom	Any symptom or description of symptoms	AO: very <mark>mild</mark> althralgias SR: very <b>also</b> arthralgias
???	When ???? (or similar) is left in the note, or when something is completely nonsense	AO: no foreign material was identified MT: no foreign ??? was identified



#### **Preliminary Results**

Our study is ongoing; final results may differ from those presented here.



#### Results

	Mean	Median	Minimum	Maximum
Length of notes (words)	558	524	102	1230
Dictation time (minutes)	5	4.5	0.4	31.5
Turnaround time – Time between completion of dictation and upload to EHR system (hours)	3.5	1	2 minutes	38.8
Clinician review time - Time between upload to EHR system and clinician signing of note (days)	4.2	1	0	42

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#### **AHR** Overall Error Rates and General Types

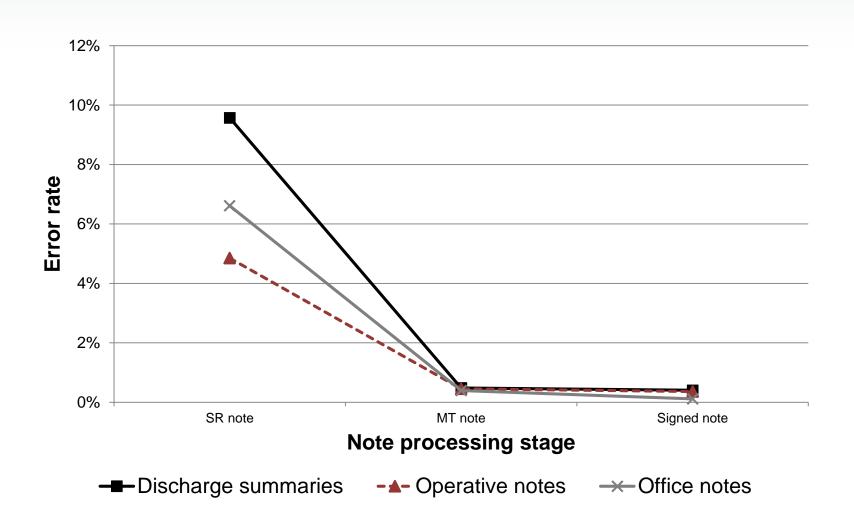
		Total Errors	Errors – General Types n (%) <sup>2</sup>		
		n (%)¹	Deletion	Insertion	Enunciation
Discharge	SR	3892 (9.6)	1395 (35.8)	1031 (26.5)	655 (16.8)
Summaries	MT	195 (0.5)	87 (44.6)	36 (18.5)	35 (18.0)
(75)	SN	163 (0.4)	74 (45.4)	29 (17.8)	29 (17.8)
Office Notes	SR	1588 (6.6)	539 (33.9)	306 (19.3)	431 (27.1)
Office Notes	MT	96 (0.4)	29 (30.2)	13 (13.5)	36 (37.5)
(55)	SN	32 (0.1)	6 (18.8)	7 (21.9)	12 (37.5)
Operative	SR	1233 (4.8)	376 (30.7)	401 (32.8)	167 (13.7)
Notes	MT	120 (0.5)	47 (39.2)	29 (24.2)	18 (15.0)
(39)	SN	96 (0.4)	42 (43.8)	25 (26.0)	15 (15.6)
All Notes	SR	6703 (7.5)	2310 (34.5)	1738 (25.9)	1253 (18.7)
All Notes	MT	411 (0.6)	163 (39.7)	78 (19.0)	89 (21.7)
(169)	SN	291 (0.3)	122 (41.9)	61 (21.0)	56 (19.2)

 $<sup>^{1}</sup>$  n = number of errors; % = total number of errors divided by the total number of words in the notes.

 $<sup>^{2}</sup>$  n = number of errors; % = number of errors of a specific type divided by the total number of errors.



### **Error Rates across All Note Types and Stages**





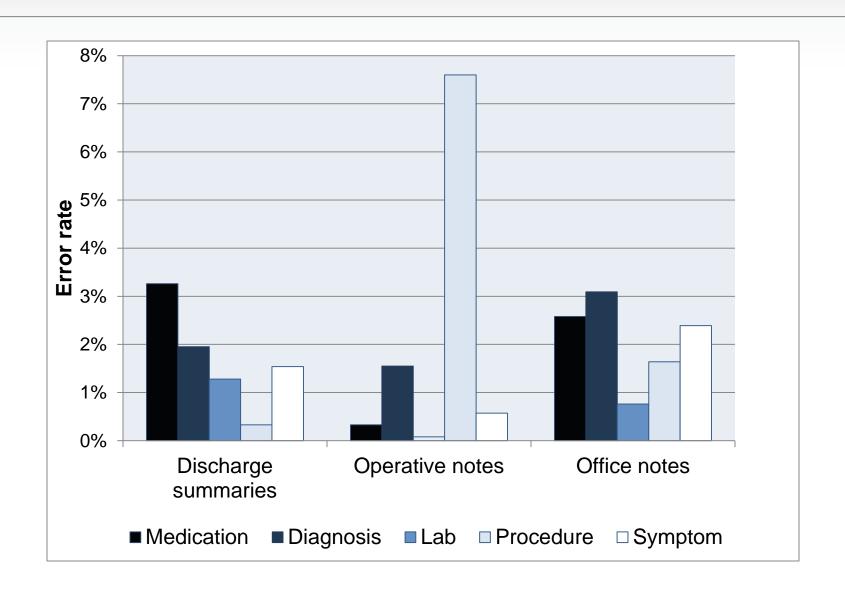
#### **Errors by Semantic Type**

			Errors – Semantic Types n (%) <sup>1</sup>						
		General	Clinical Information						
		English	Medication	Diagnosis	Procedure	Symptom	Lab	Physical Exam	Imaging
Discharge	SR	3255 (83.6)	127 (3.3)	76 (2.0)	13 (0.3)	60 (1.5)	50 (1.3)	30 (0.8)	25 (0.6)
Summaries	MT	124 (63.6)	14 (7.2)	9 (4.6)	2 (1.0)	7 (3.6)	2 (1.0)	14 (7.2)	2 (1.0)
(75)	SN	105 (64.4)	6 (3.7)	9 (5.5)	2 (1.2)	7 (4.3)	2 (1.2)	14 (8.6)	2 (1.2)
Office	SR	1305 (82.2)	41 (2.6)	49 (3.1)	26 (1.6)	38 (2.4)	12 (0.8)	7 (0.4)	7 (0.4)
Notes	MT	79 (84.1)	1 (1.0)	4 (4.2)	2 (2.1)	2 (2.1)	1 (1.0)	0 (0.0)	0 (0.0)
(55)	SN	28 (87.5)	0 (0.0)	2 (6.3)	0 (0.0)	1 (3.1)	1 (3.1)	0 (0.0)	0 (0.0)
Operative	SR	947 (77.4)	4 (0.3)	19 (1.6)	93 (7.6)	7 (0.6)	1 (0.1)	4 (0.3)	1 (0.1)
Notes	MT	81 (67.5)	0 (0.0)	4 (3.3)	15 (12.5)	2 (1.7)	0 (0.0)	0 (0.0)	1 (0.8)
(39)	SN	72 (75.0)	0 (0.0)	3 (3.1)	9 (9.4)	0 (0.0)	0 (0.0)	0 (0.0)	1 (1.0)
All No.	SR	5507 (82.2)	172 (2.6)	144 (2.2)	132 (2.0)	105 (1.6)	63 (0.9)	41 (0.6)	33 (0.5)
All Notes (169)	MT	284 (69.1)	15 (3.7)	17 (4.1)	19 (4.6)	12 (2.9)	3 (0.7)	15 (3.7)	3 (0.7)
(169)	SN	205 (70.4)	6 (2.1)	14 (4.8)	11 (3.8)	9 (3.1)	3 (1.0)	15 (5.2)	3 (1.0)

<sup>&</sup>lt;sup>1</sup> n = number of errors; % = number of errors of a specific type divided by the total number of errors



#### **AHRO** Clinical Information Errors in SR Notes





### **Clinical Information Errors Across Note Stages**

	Total Errors	Clinical Information Errors n (%)	General English Errors n (%)	Other Errors n (%)
SR	6703	691 (10.3)	5507 (82.2)	505 (7.5)
МТ	411	84 (20.4)	284 (69.1)	43 (10.5)
SN	291	61 (21.0)	205 (70.44)	25 (8.6)

Other errors include patient and provider information, dates and ???.

 40% of SR notes, 7% of MT notes, and 5% of SN notes contain at least one clinically significant error.



### Content Rearranging and Stylistic Changes

	Medical Transcriptionist	Clinician
Rearranged text	17.4%	6.4%
Made stylistic changes	91.7%	43.1%
Added information	N/A	29.7%
Deleted information	N/A	24.8%

<sup>% =</sup> number of notes where changes were made divided by total number of notes.



#### **Error Examples**

- TB vaccine vs. TD vaccine.
- Staining vs. standing.
  - "Continues to have daily standing" (pt's menstruation).
  - SR, transcriptionist, and signed note all missed this error.
- Menorrhagia and gluten allergy were missed by SR and transcriptionist, and remained omitted on the signed note.
- SR and transcriptionist missed the name of the drug and listed as ??\_\_??
  - ► The drug was *celecoxib*. The SR and transcriptionist notes did not record it. The signed note listed drug as *naproxen*!



# Discussion: Productivity and Quality

- Back-end dictation service had a relatively quick turnaround time and a low error rate.
- While many errors were generated by SR, most (~94%) were corrected by the medical transcriptionist manually.
  - The addition of a human editing an SR-generated note is invaluable.
- 20% of EHR-related malpractice cases were due to incorrect information in the EHR.<sup>1</sup>
  - Without the MT revising the notes, clinically significant errors could have had a negative impact on patient care and potentially caused legal issues.

<sup>&</sup>lt;sup>1</sup>Ruder DB. Malpractice claims analysis confirms risks in EHRs. Patient Safety & Quality Healthcare. Jan/Feb 2014: Volume 11, Issue 1.



#### **Discussion: Error Checking**

- Errors still left in the signed note suggest that some providers may not review their dictated notes thoroughly or at all.
  - ▶ 7% of signed notes contained a blank space the transcriptionist marked as ?? ??
- If physicians use SR directly, they may have to spend a considerable amount of time correcting the SR-generated text.
  - Although the errors might be less than our results, since the SR can be trained by the individual physician.
- Automated error detection may help improve the accuracy of dictated documents.



#### **Ongoing and Next Steps**

- Conduct error analysis for front-end user of SR.
  - Clinical observations
  - Simulations
- Build a knowledge base.
  - Confusion sets
  - Error frequencies
  - Error patterns
- Develop automated methods to detect SR errors.
  - Statistical methods (noisy channel models, co-occurrence statistics), machine learning, and knowledge-based methods.



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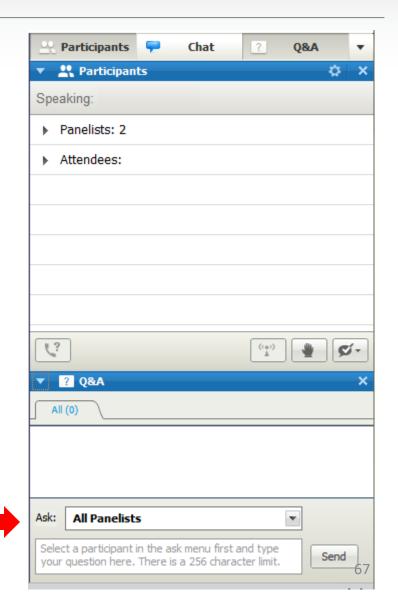
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